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A class within a four classes module



- University of Verona,
- School of Exercise and Sport Science,
- Laurea magistrale in Scienze motorie preventive ed adattate

Metodologia delle misure delle attività sportive

Wednesday 30/10/2013



Accelerometer issues - need for custom developed software...



measures

- PROPRIETARY ALGORITHM (i.e., 'how from count to ME?');







Accelerometer Accelerometer issues - From linear to non-linear ME=f(counts) -> 3D accelerom. -50->-3% nME

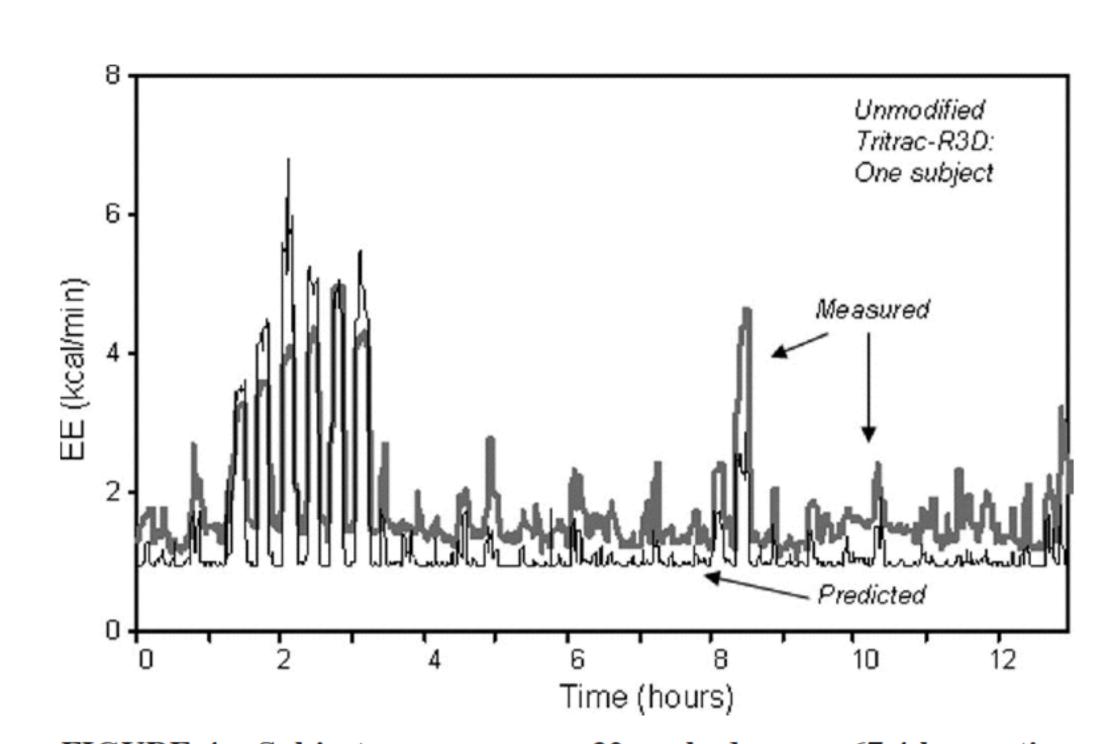
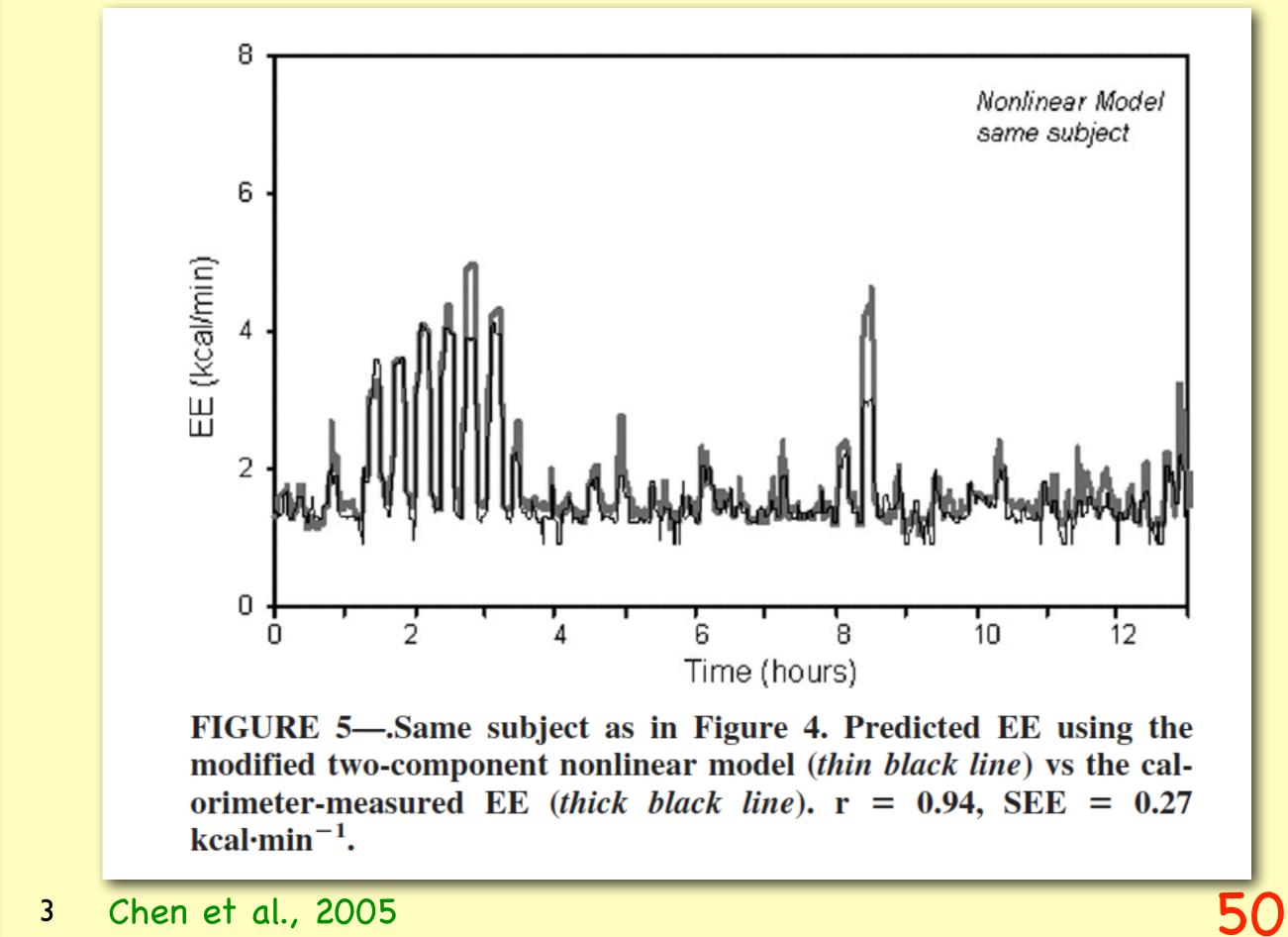


FIGURE 4—Subject: a woman age 32 yr, body mass 67.4 kg, resting $EE = 1.06 \text{ kcal} \cdot \text{min}^{-1}$. Tritrac-predicted EE (*thin black line*) vs the calorimeter-measured EE (thick black line) during the waking period of a 24-h stay in the room calorimeter. r = 0.88, SEE = 0.48 kcal·min^{−1}.

measures









-> TriTrac-R3D -> RT3

ActiGraph 71-64/-256

-> ActiGraph GT3X





measures

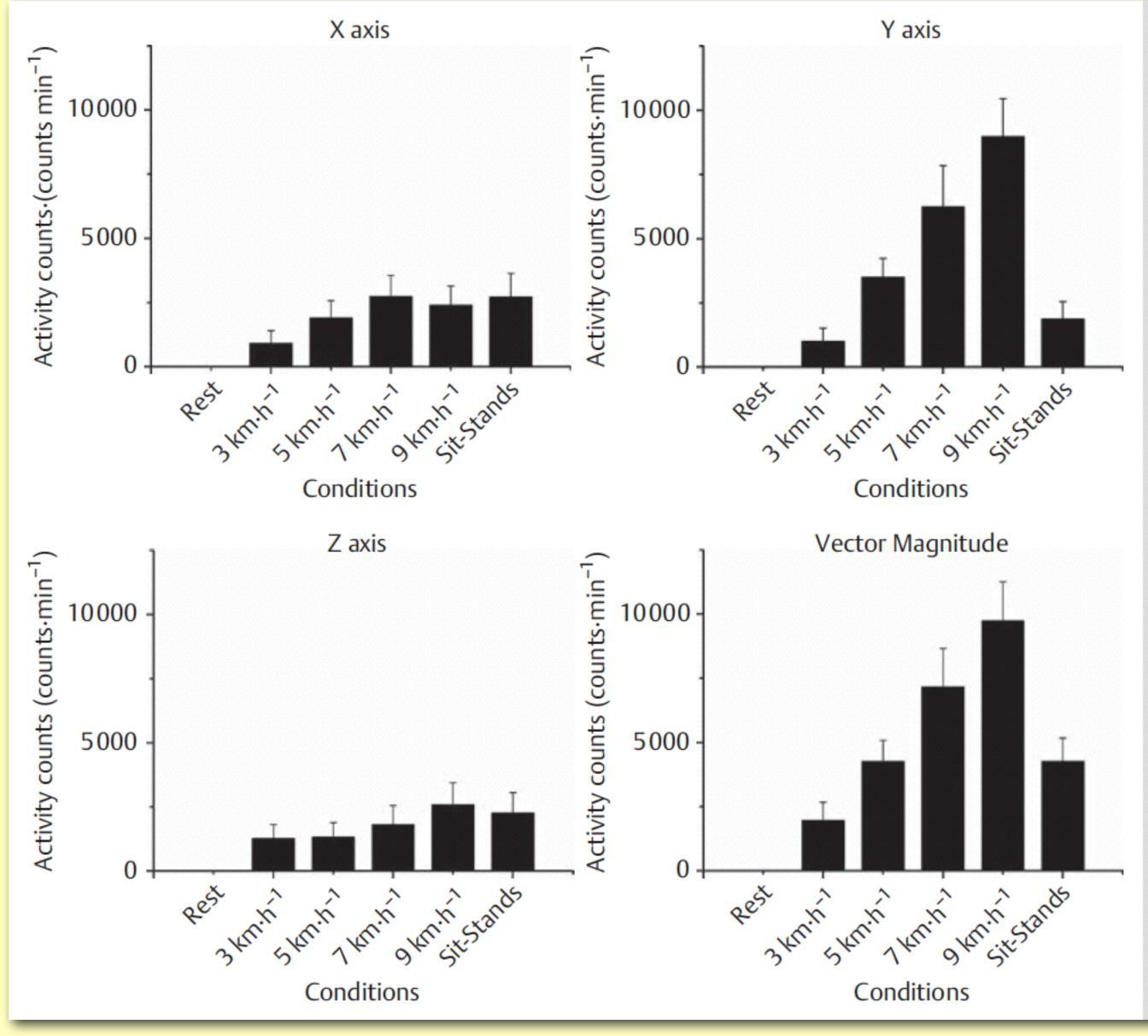


ActiGraph GT1M ->





->

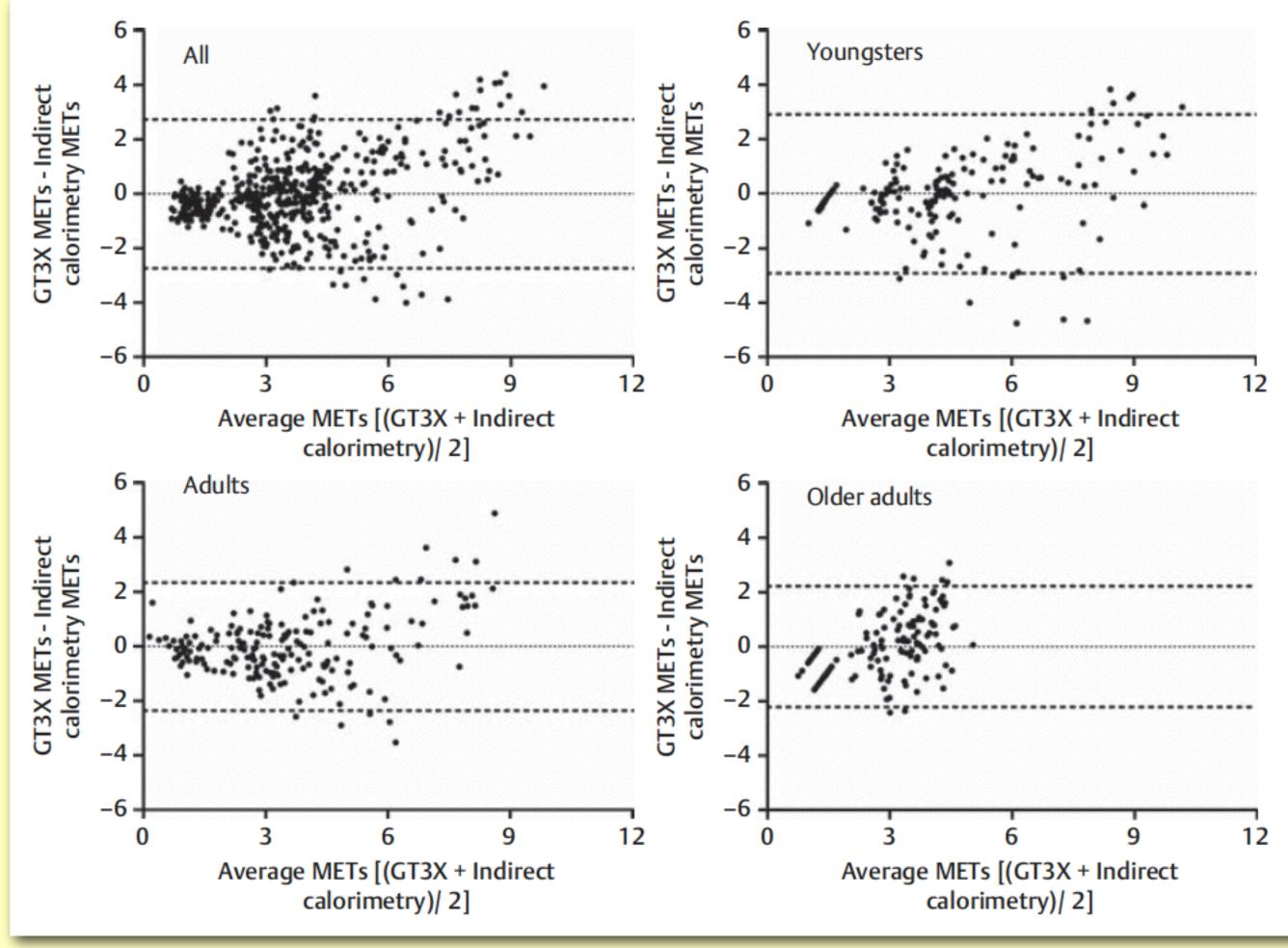


Santos-Lozano et al., 2013

measures

Fig. 1 Activity counts (counts \cdot min⁻¹) (mean ± standard deviation) per axis and activities for all participants.





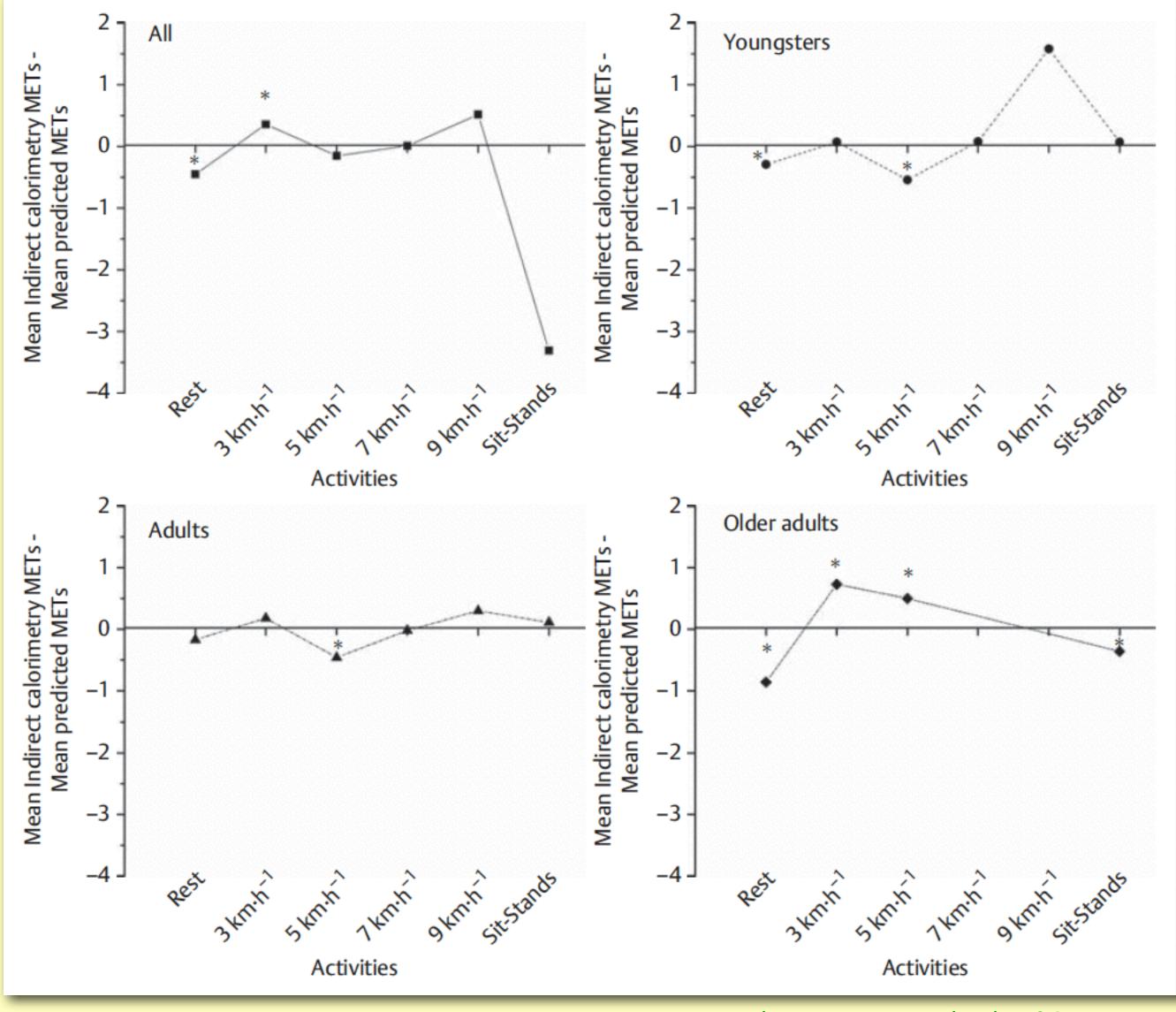
measures

Fig. 3 Bland and Altman Plots in each group (energy expenditure (EE, in METs) determined with indirect calorimetry – EE (METs) predicted with GT3X).

Santos-Lozano et al., 2013







Santos-Lozano et al., 2013

measures

Fig. 4 Energy expenditure (EE, in METs) from indirect calorimetry vs. EE predicted with the GT3X for each age-group. *Significantly different from indirect calorimetry vs. predicted, same activity and age-group, *P*<0.05.



Actiwatch



-> Actical



Actitrac



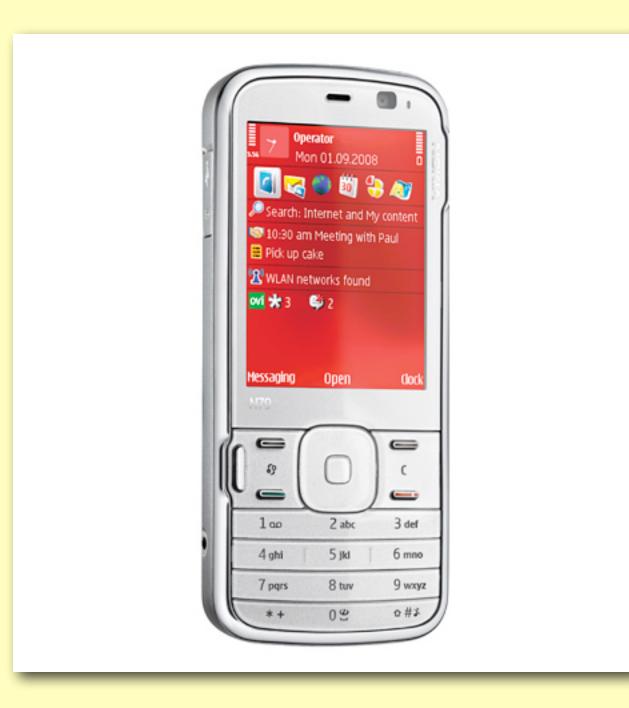
Biotrainer

measures





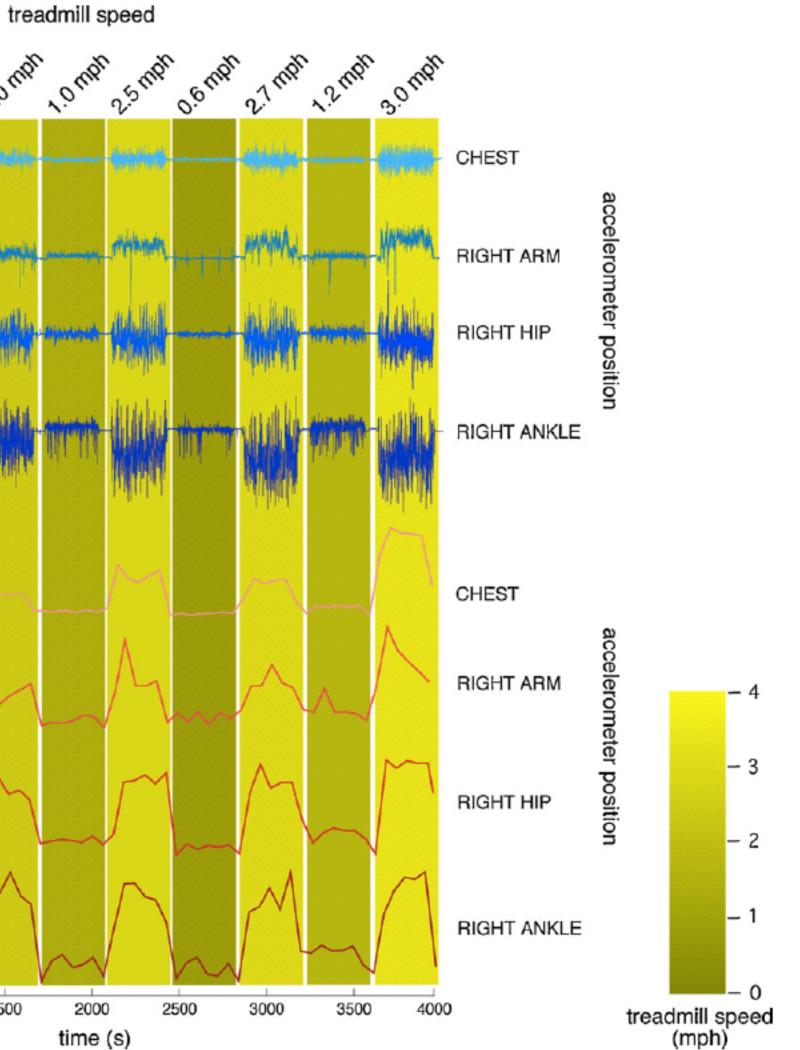
Nokia N79



measures

Carlson Jr et al., 2012





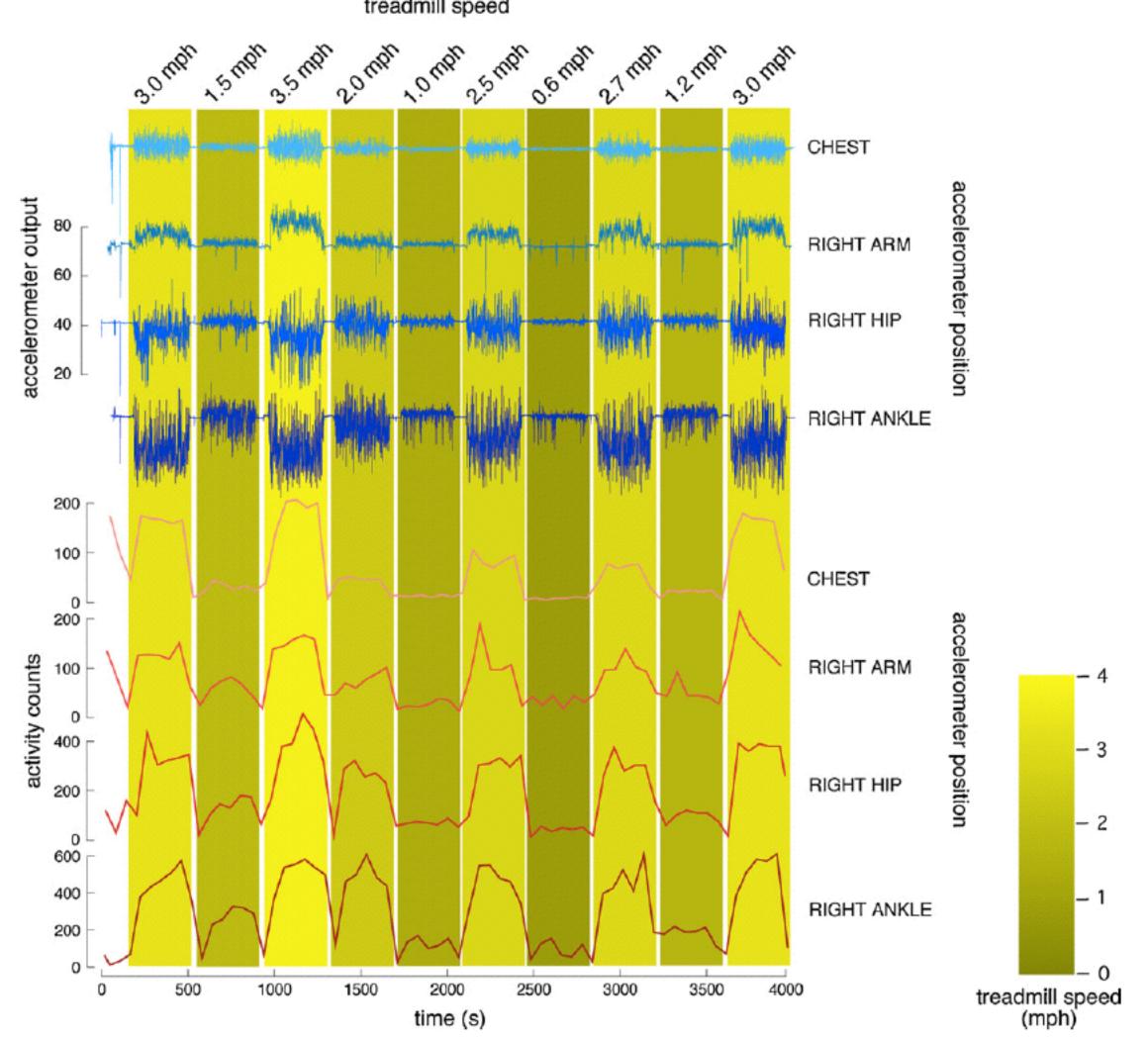
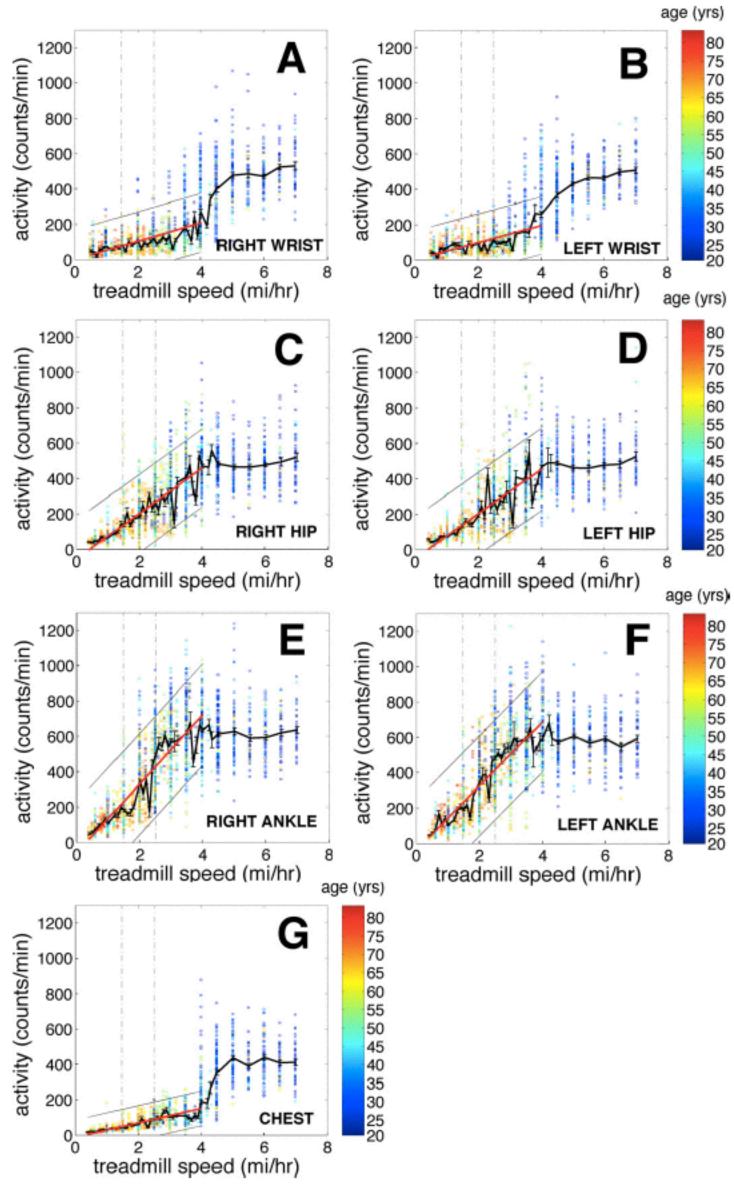


Fig. 1. Activity counts from cell-phone accelerometers provide an accurate measure of treadmill gait speed regardless of where the sensor is worn. The top four traces depict raw data from a representative trial (43 y/o man) showing acceleration magnitude versus time for sensors worn at the chest, right arm, right hip, and right ankle (1st through 4th traces from top, respectively). For all traces the baseline is centered at 64 (midscale between sensor output of 0 for -2 g, and 128 for +2 g), the amount of deflection from this baseline is per the common scale provided left of these traces. The bottom four traces show activity counts versus time for the sensors worn at the chest, right arm, right hip, and right ankle, respectively. Counts were calculated over 1 min nonoverlapping bins. Treadmill speed is given at the top of each epoch bar.





(A), left wrist (B), right hip (C), left hip (D), right ankle (E), left ankle (F), and neck (G).

Accelerometers

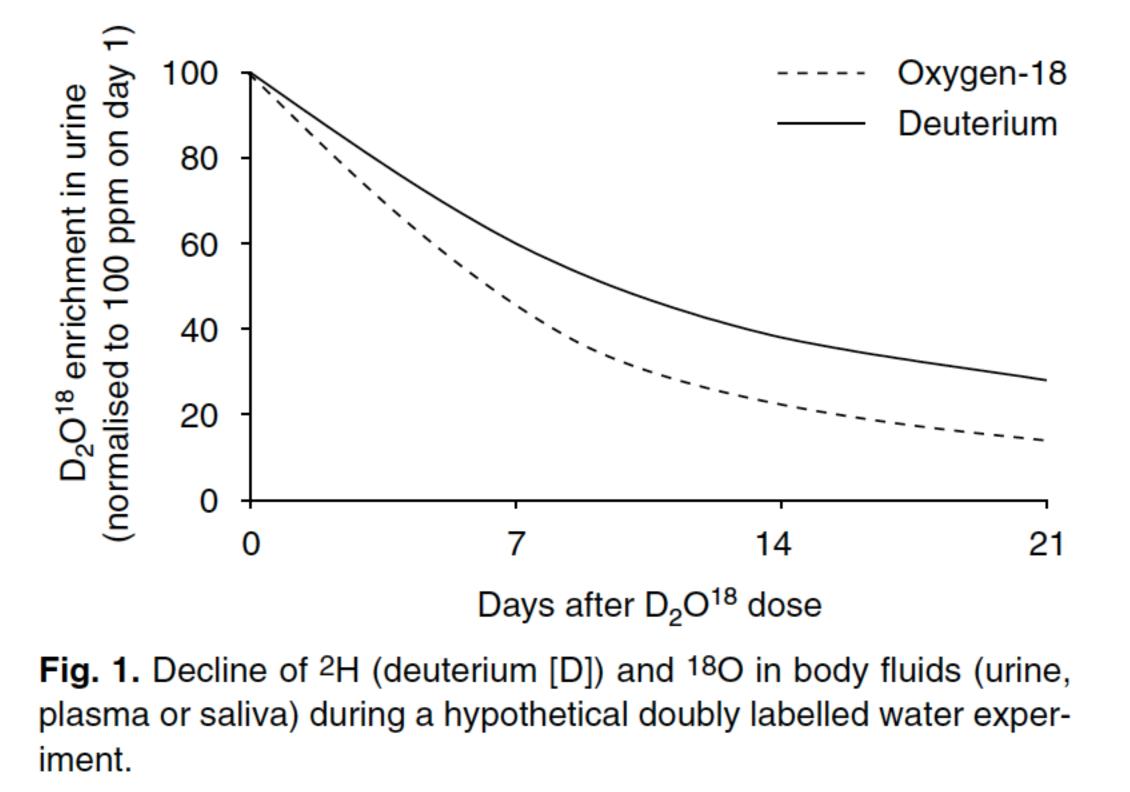
Fig. 3. Activity count versus treadmill speed relationships for all sensor locations. For all figures, the solid red line shows the linear regression between treadmill speed and activity counts (fit for all data between 0.0 and 6.4 km/h (0-4 mi/h) gait speeds); the thin surrounding black lines are 95% confidence boundaries on this regression. The thick black line connects mean activity count values for each of the evaluated treadmill speeds; bars surrounding this point are ±1 standard error of the mean. Individual observations of activity counts are shown as open colored circles. Subject age is color coded as circle color; refer to colorbar at right side for key. The dashed lines at gait speeds of 2,35 km/h (1,46 mi/h) and 4 km/h (2,5 mi/h) highlight system performance at two critical functional thresholds. These relationships come from cell phones placed at the right wrist





- Lifson et al., 1955;
- (small animals) 1975;
- validation by Scholler et al., 1982;
- (premature infants, children, pregnant and lactating women, elderly, obese people, hospitalized patients);
- subject is administered a dose of stable isotope ²H₂¹⁸O, which (²H, ¹⁸O) equilibrates relatively quickly with body water (H, O);
- ²H is eliminated as ²H₂O (breath, urine, sweat, perspiratio insensibilis), while the ¹⁸O is eliminated either as $H_2^{18}O$ (breath, ...) and as $C^{18}O_2$ (breathe only);
- difference between the two rates of elimination -> V'CO2 -> ME





measures



- RQ (= V'CO2 / V'O2) estimate -> reliability: . standard Western diet -> RQ estimate; . food intake diary -> RQ estimate (i.e., food quotient \approx RQ); . indirect calorimetry -> RQ



DLW method issues

- intensity of each type) to ME;
- have considerable costs;
- -> only 3-4 ÷ 21 d ME;
- unknown RQ -> 5% e

measures

- inability to discriminate the contribution of individual PAs (types, amount,

- costs: isotopes and tools to detect them (i.e., mass spectrophotometers) still







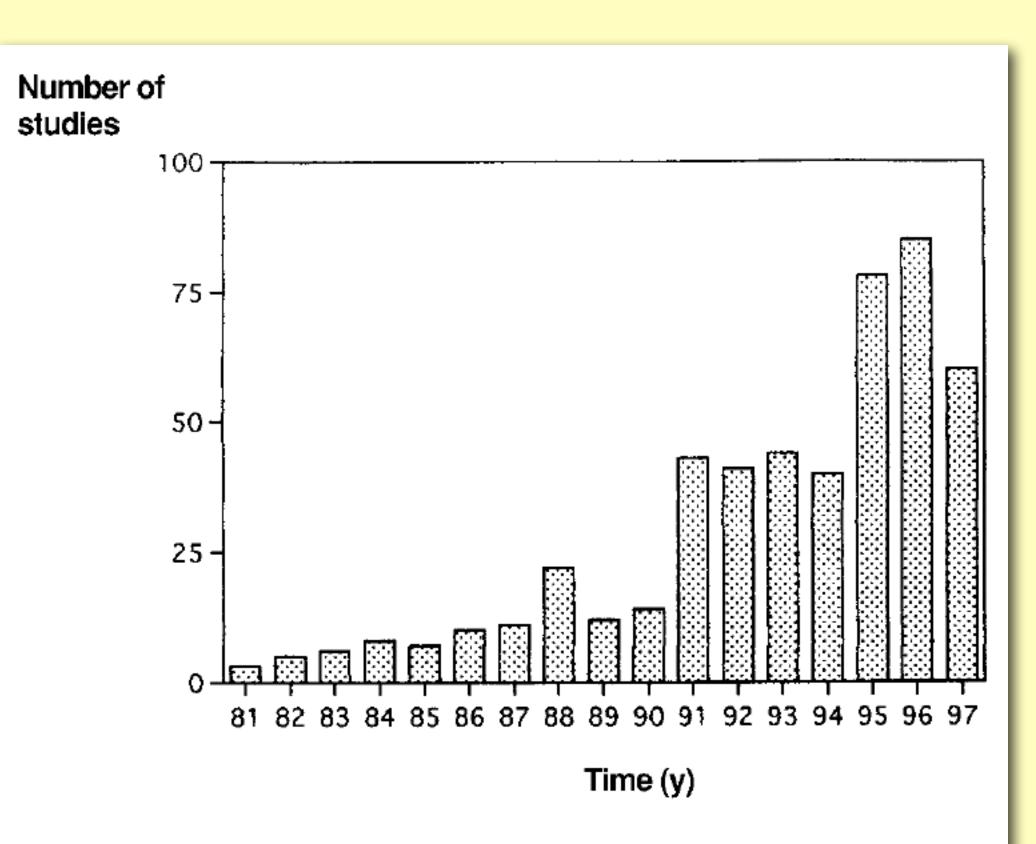


FIGURE 1. Number of studies in peer-reviewed journals (excluding abstracts) that used the doubly labeled water technique in the years 1981–1997 (through June) from the Science Citation Index (Institute for Scientific Information, University of Aukland, New Zealand). Since the first study in humans in 1982 the use of the technique has continued to grow.

measures





Accelerometer issues

- SINGLE-SITE PLACEMENT;
- speed rapid changes activities (e.g., tennis)

measures

- waist placement -> PA underestimate during upper limb movement, standing, vertical activity (i.e., climbing stairs, uphill walking), pushing or pulling objects, carrying loads (e.g., books or laptops), body-supported exercise (e.g., cycling), water PA (e.g., swimming), running faster than 9 km/h, horizontal



Solution?

- A combination of variables describing: movements feature sedentary PA); 2) a trunk-focused posture variable featuring locomotion; 3) lower limbs-focused high intensity components (lower limbs have largest, most powerful muscles);

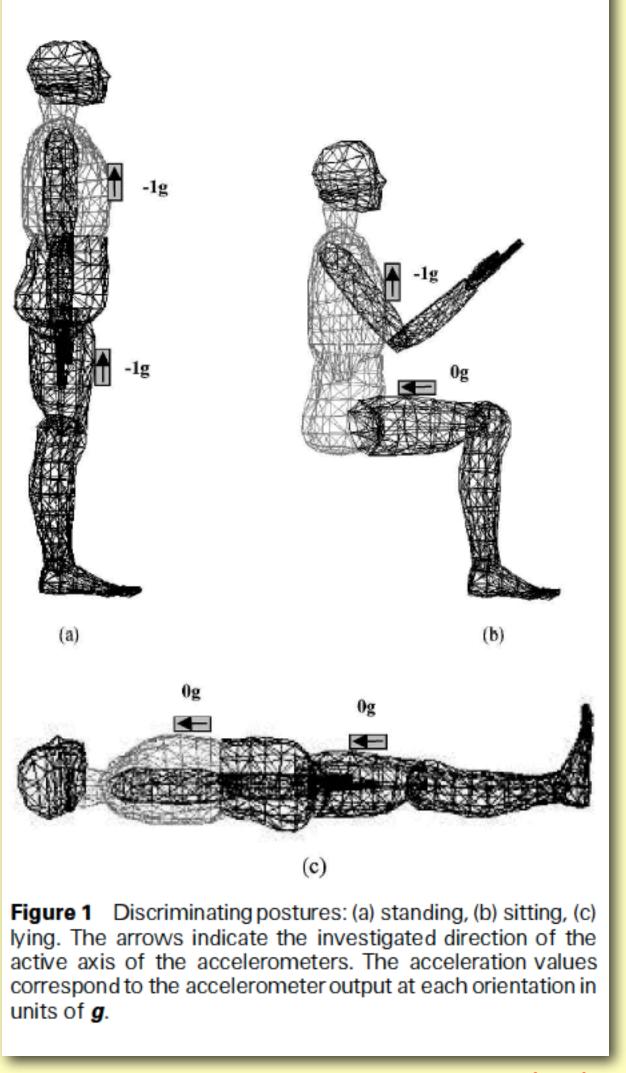
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1) upper limbs-focused high frequency components (upper limbs
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- More than ONE accelerometer together, as well (e.g., waist TriTrac-R3D + dominant arm wrist Actiwatch, Actiwatch + Actical, ...);
- accelerometers based activity logger: . two (@sternum, front thigh) biaxial accelerometers + analog data-logger;

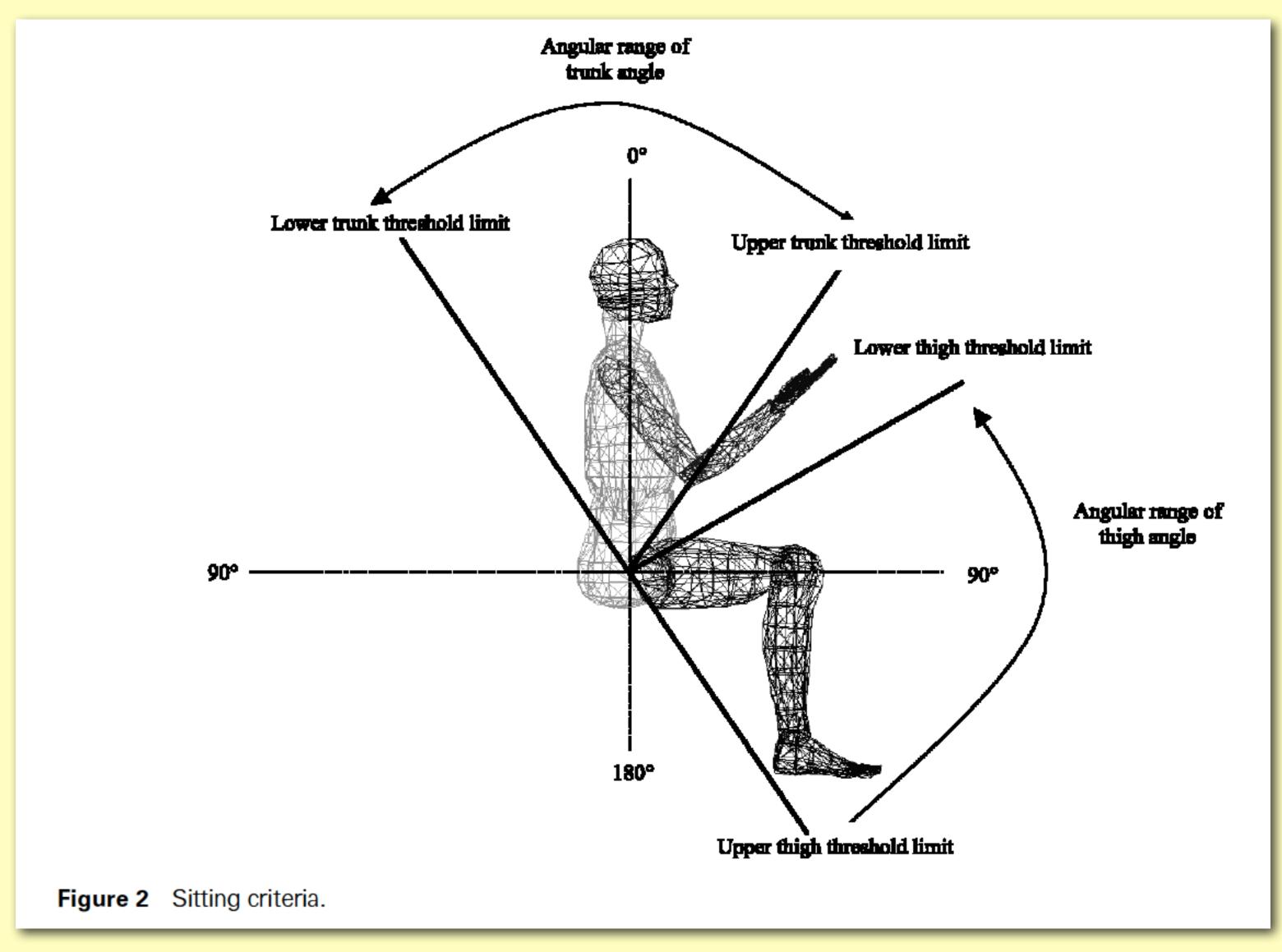
measures



Culhane et al., 2004 66







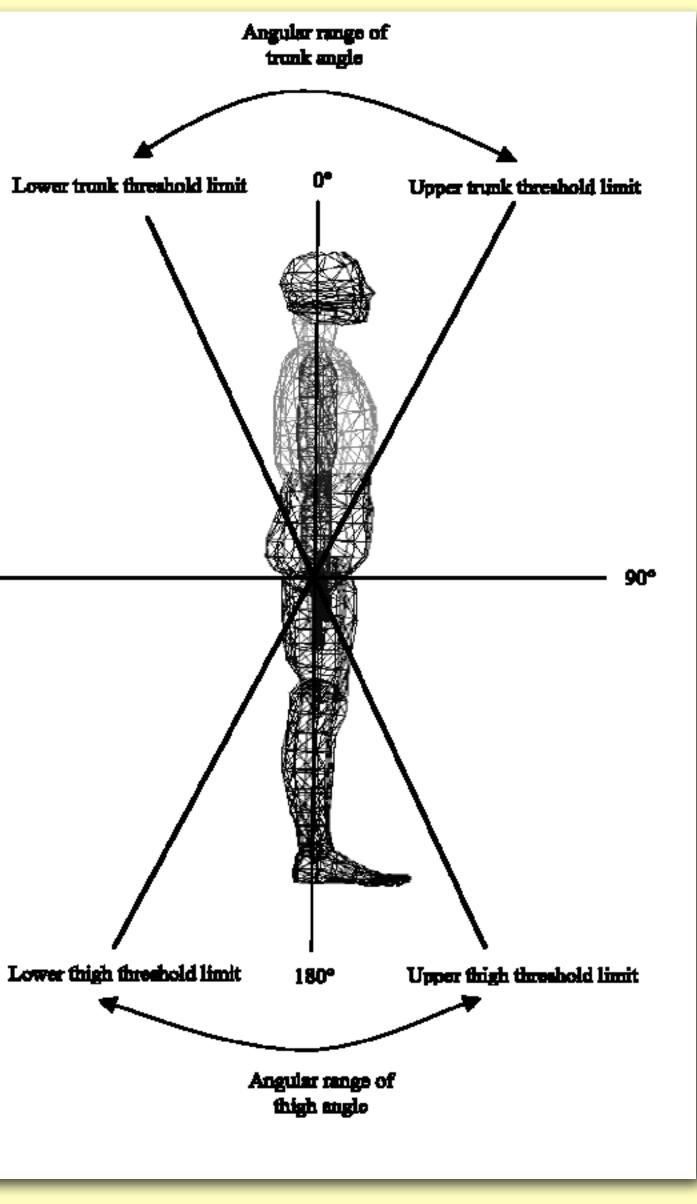
Culhane et al., 2004

measures



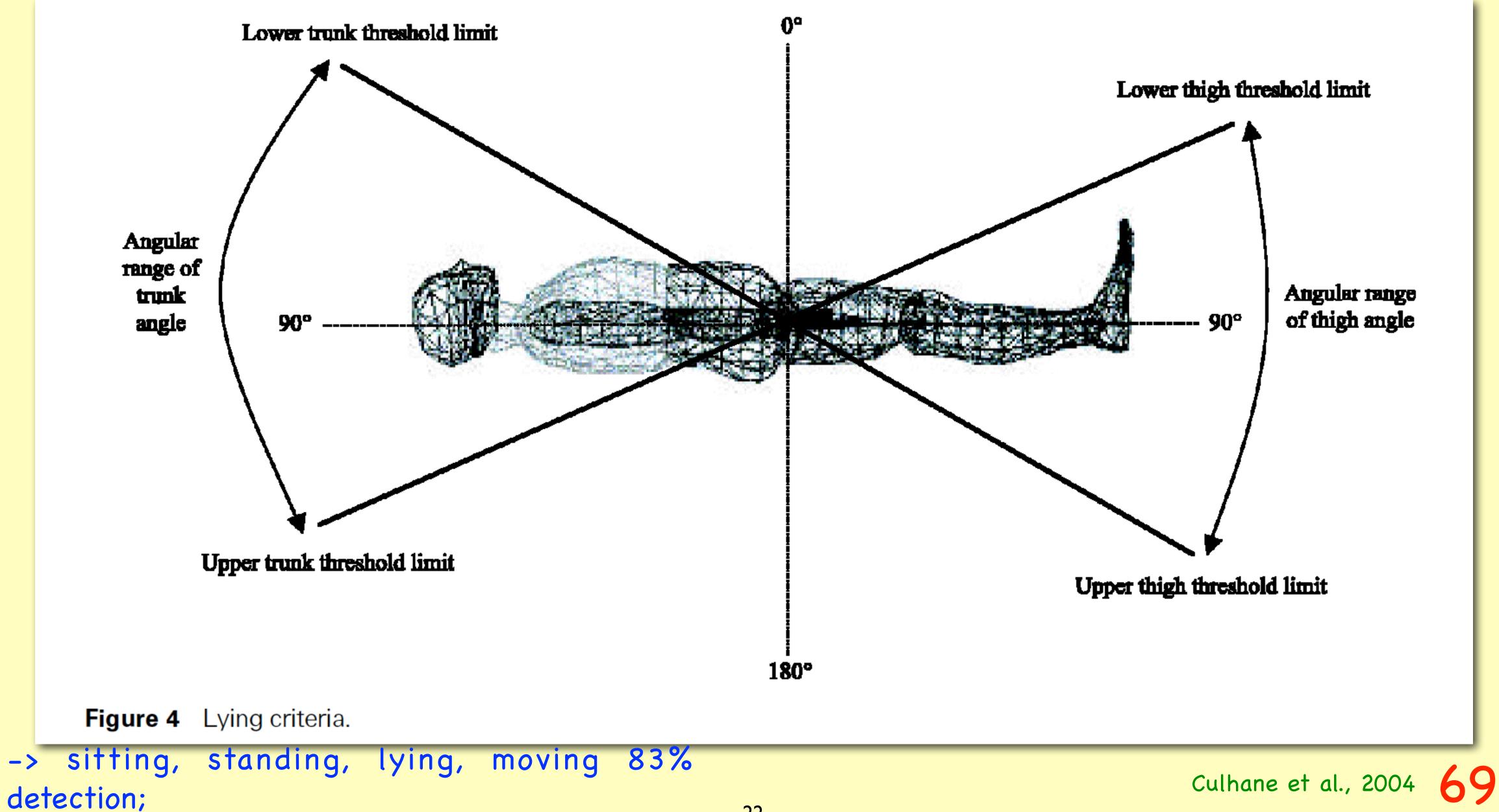
90*

Figure 3 Standing criteria.



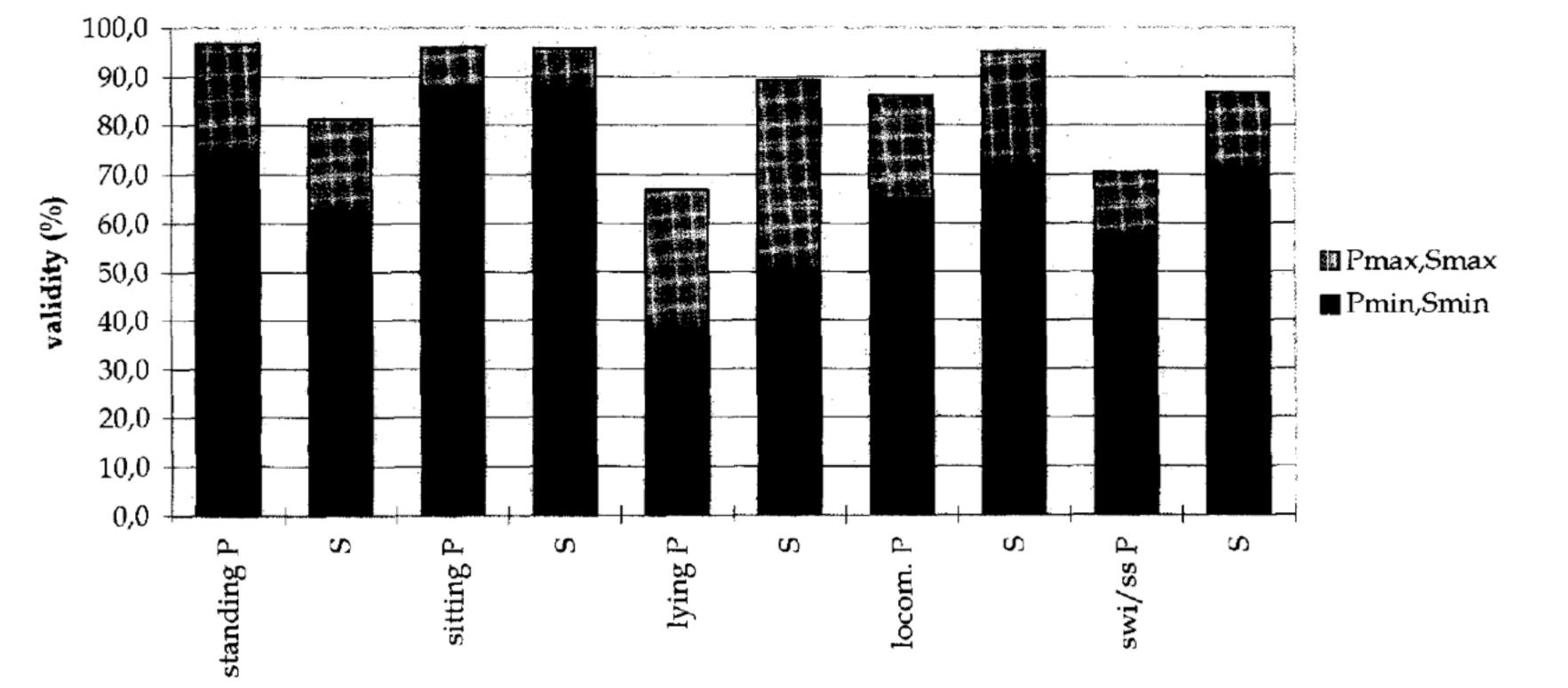








min. and max. predictive value and sensitivity per class



Busser et al., 1997 . uniaxial accelerometer (@front thigh) + 2 unixial accelerometer/digital data-logger (backpack) -> sitting, standing, lying, crawling, walking, running, 23 going on a swing 73+91% detection;

measures

class

Figure 6 Minimal and maximal validity of the individual ADL categories based on the monitor's sensitivity (S_{min} and S_{max} , respectively) and predictive value (P_{\min} and P_{\max} , respectively). Sensitivity indicates how often the monitor recognizes a category; the predictive value indicates how often the decision of the monitor is correct. A lack of sensitivity indicates a false negative; a lack of predictive value indicates a false positive.



. three uniaxial accelerometers (2@sternum, front thigh) + digital recorder;

-> sitting, standing, lying, walking, climbing/going down stairs, cycling 80% detection (Veltink et al., 1996);

. four biaxial accelerometers (@lateral thighs, sternum or front forearms) + HR monitor + digital recorder;

-> more than twenty different postures/locomotions 83÷88% detection;

measures

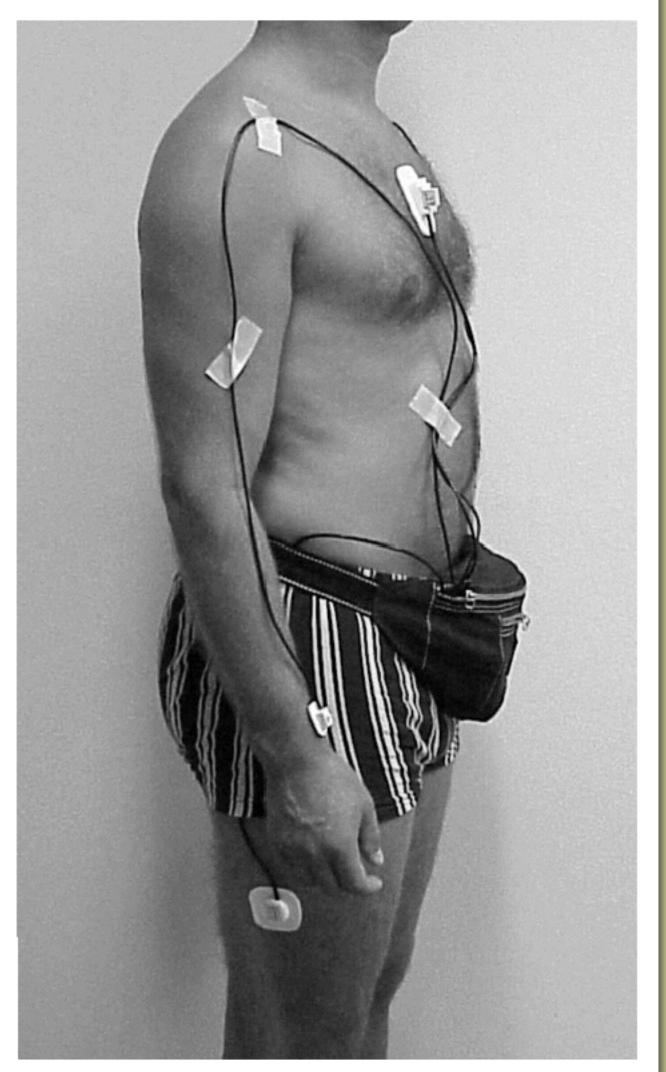


Figure 1. An extended configuration of the Activity Monitor, with accelerometers at the thighs, trunk, and lower arms.

Bussmann et al., 2001



- Introduction of another type of physical sensor:

. (@sternum) two biaxial accelerometers + piezoelectric gyroscope + digital recorder (@wrist);

measures

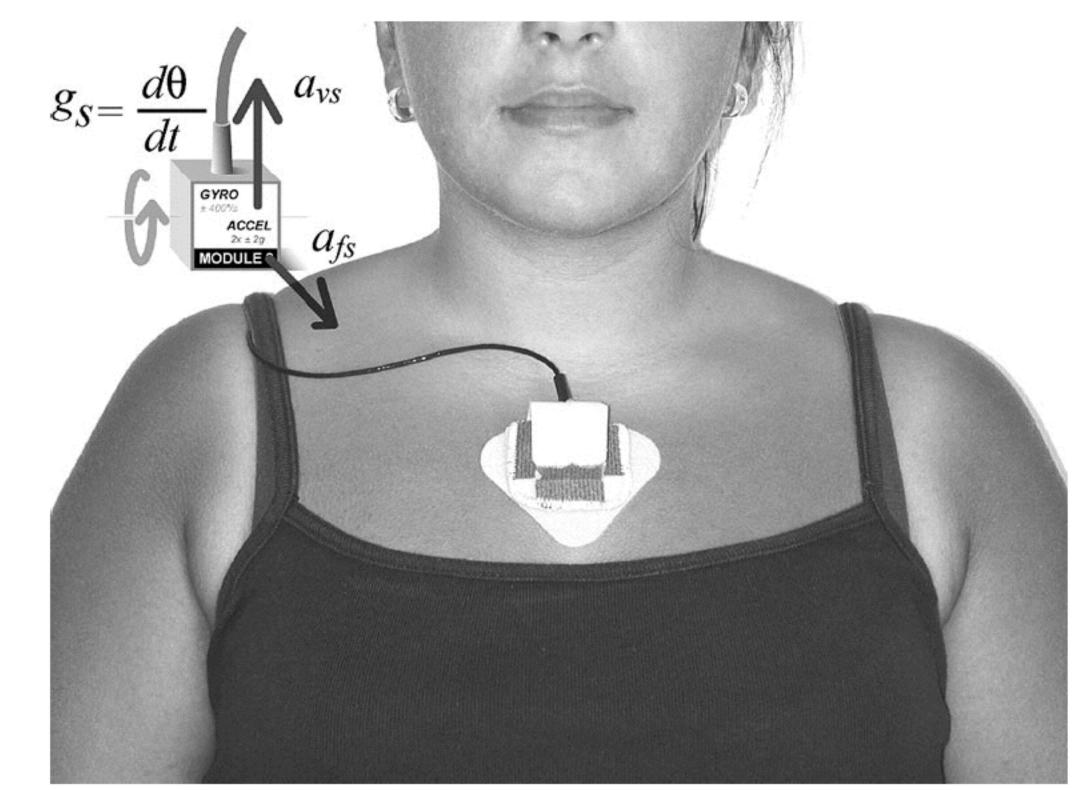


Fig. 1. Sensor attachment. Vertical and frontal acceleration $(a_{vs} \text{ and } a_{fs})$ as well as angular velocity (g_s) are measured using a kinematic sensor attached to the subject's chest.

Najafi et al., 2003







TABLE II Overall Sensitivity and Specificity of Transition Detection for the 11 Elderly (First Study)								
# Test	Total PT [*]	Sensitivity, %					Specificity, %	
		ΡT	SiSt ^{**}	StSi	Lying	Walking	SiSt	StSi
1	40	100	100	100	100	95±4	100	100
2	66	98±5	100	97±10	-	97±3	95±12	100±0
3	58	100	97±10	63±29	-	-	63±29	97±10
4	58	100	88±25	75±29	-	-	75±29	88±25
5	64	96±9	89±18	86±19	-	-	86±19	94±13
6	57	100	85±19	72±24	-	-	72±24	85±19
Mean	57±9	99±2	93 ±7	82±15	100	96±1	82±15	94±6

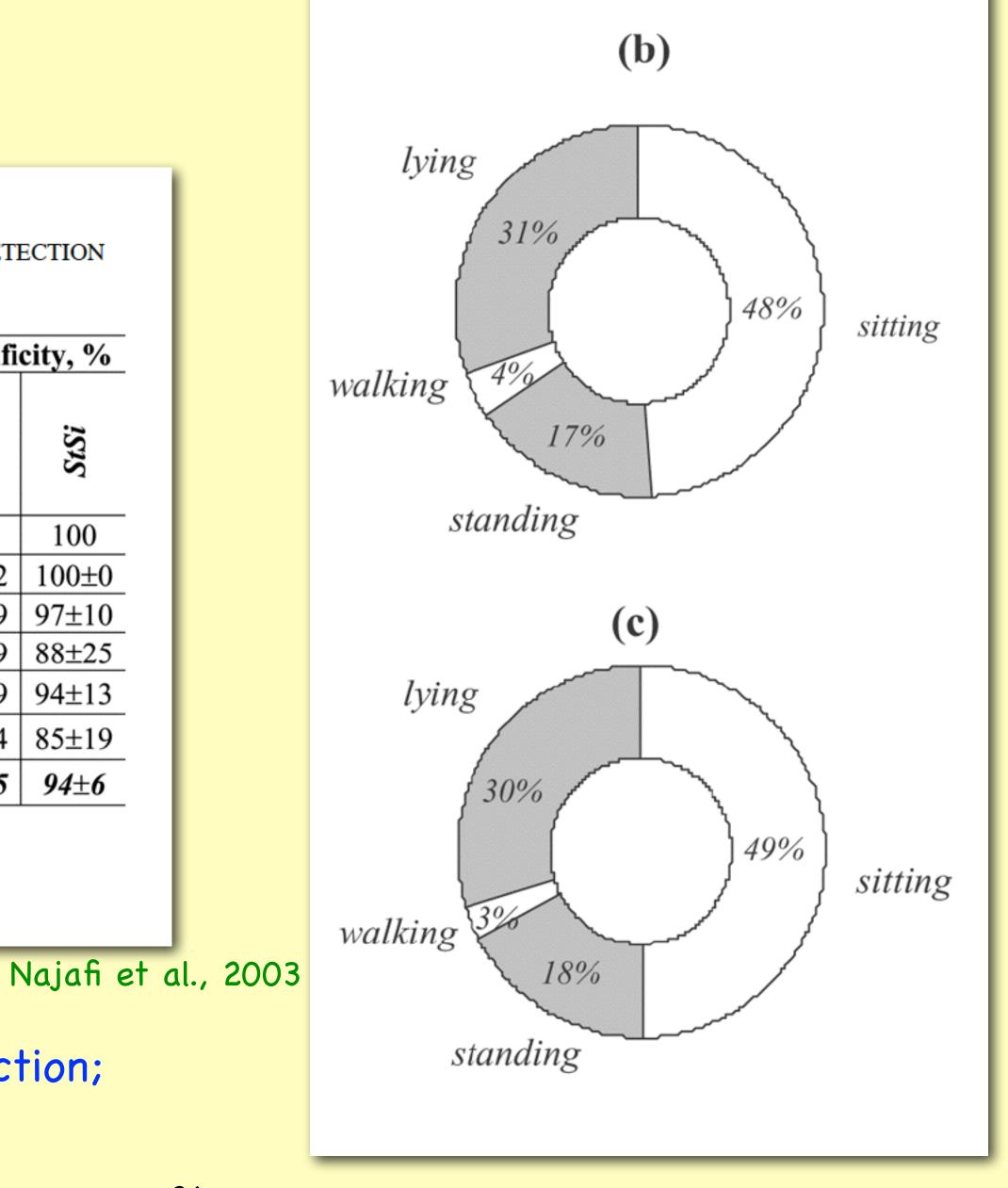
* PT: Postural transition.

** SiSt: sit-to-stand transition.

† StSi: stand-to-sit transition.

-> posture change, walking detection;

measures







- thermometry, ventilation measure):
 - . e.g., HR monitor (-> ME) + motion sensor(s) (-> motion-sensor-sensitive PA);
- accelerometers + inclinometers -> body position over time -> 85% unstructured exercise thermogenesis estimate:
 - . total internal heat produced $\approx 75 \div 80\%$ energy intake;
 - . partial internal heat produced <- sitting, standing, walking, working, any other unstructured exercise;

 - . i.e., motion sensor -> yes/not time to use HR monitor for ME estimate;

measures

- Accelerometry (-> movement) + physiological measure (e.g., HR measure,

. proposal: (during the day) wearing motion sensor, (structured exercise) wearing HR monitor;



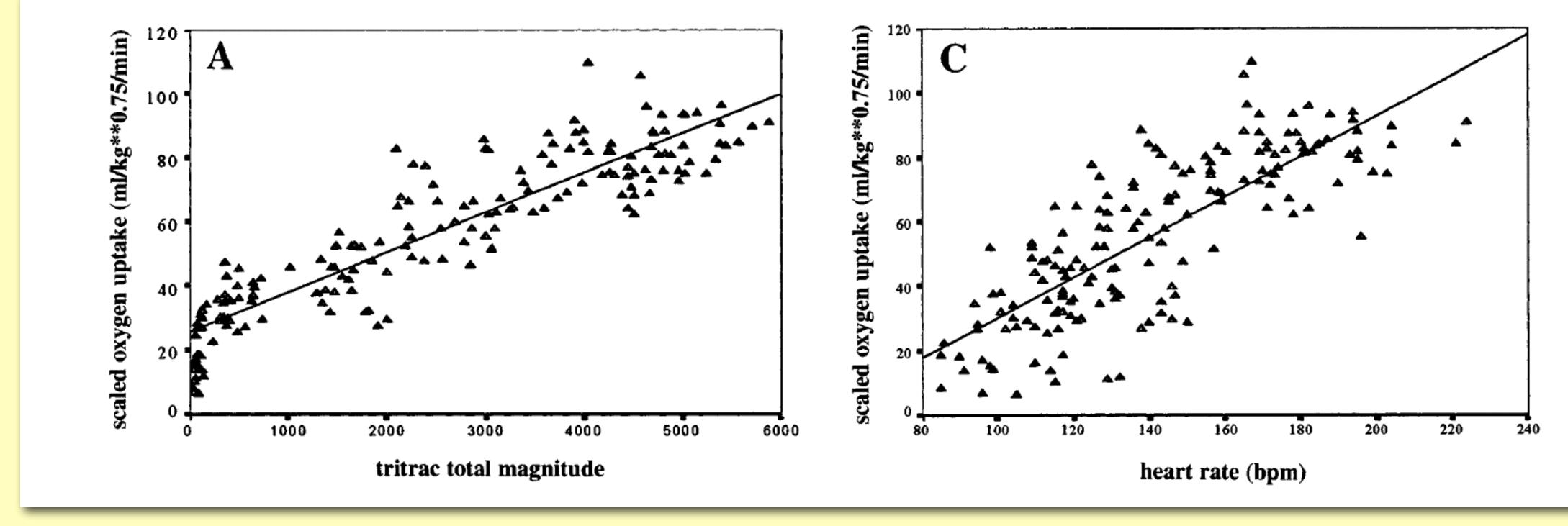












. exception: children (i.e., V'O2 [ml O2/kg^{.75} min] correlated w/both counts, HR, but w/counts r² > w/HR r²);

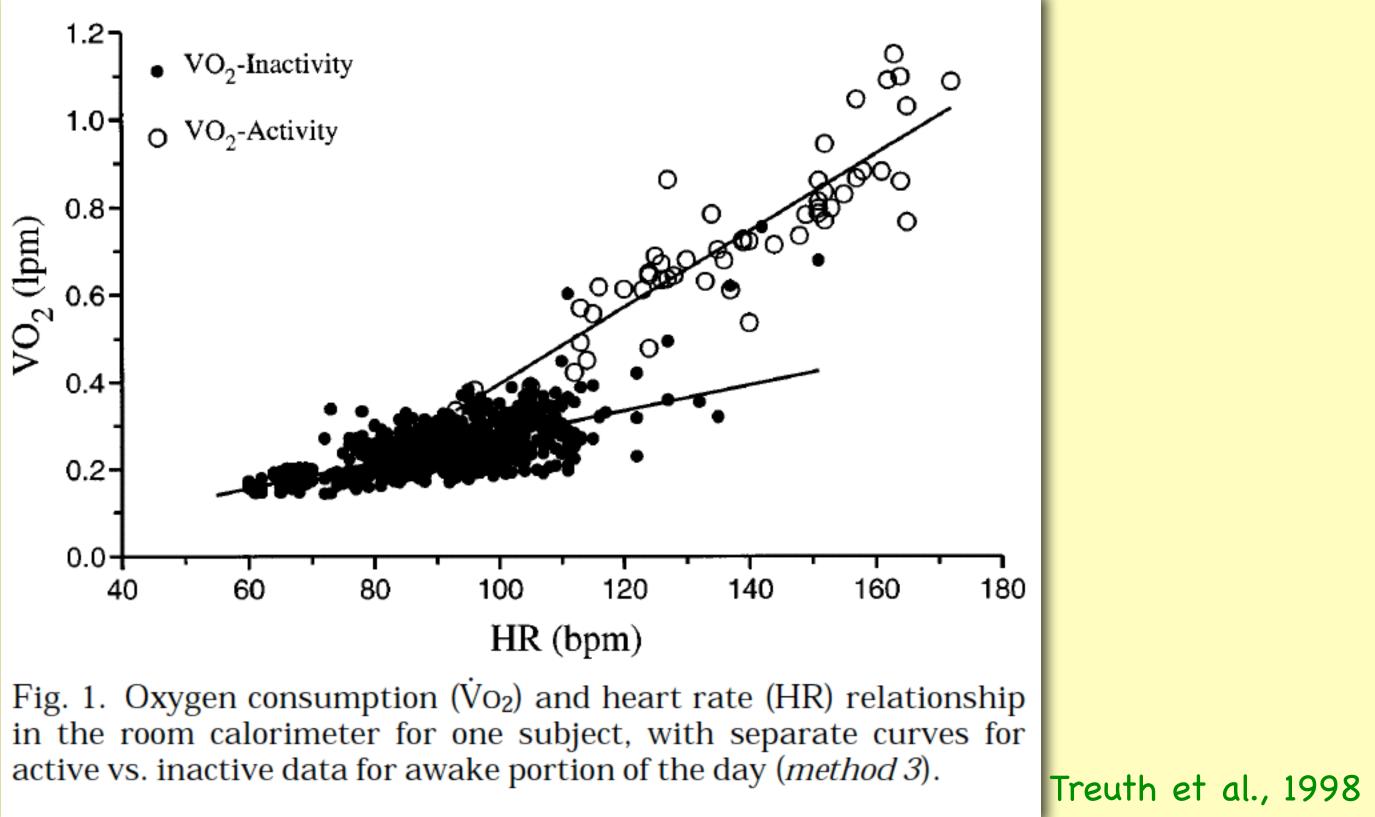
measures

Eston et al., 1998





Second generation accelerometers (re: children HR)



measures

. solution: two different individual V'O2 vs. HR relationships, one for inactivity, one for PA;



- Accelerometry + HR measure:

- . FitSense FS-1;
- . Actiheart:
 - @chest;
 - each subject's calibration;
 - OPEN ALGORITHM;
 - user's models;
 - accelerometer-, HR monitor-, accelerometer+HR

```
monitor-driven model;
```













. SenseWear Armband:

- accelerometer + heat flow sensor (-> `internal heat produced') + skin galvanic response sensor (-> evaporation heat loss) + skin thermometer + instrument's shell (i.e., near-body) thermometer;
- gender, age, height, mass input;
- PROPRIETARY ALGORITHM (I.E., 'HOW FROM EACH SENSOR'S OUTPUT TO ME?');
- -> -18÷-7% walking, stairs climbing, cycling V'O2 ME; -> -29% armergometer V'O2 ME;
- <- investigators results driven new PROPRIETARY</p> algorithm developed -> n.s. differences; -> underestimate of rowing V'O2 ME; arm cutaneous fat issue;

-> good precision of resting V'O2 ME; -> good precision/low accuracy of cycloergometer V'O2 ME;

measures











- \rightarrow +13÷+27% level walking V'O2 ME;
- -> -22% uphill walking V'O2 ME;
- -> overestimate of walking, running V'O2 ME;
- -> overestimate of wheelchair users activities V'O2 ME;
- -> underestimate of obese subjects resting V'O2 ME;
- -> overestimate of obese subjects exercise V'O2 ME;
- -> good accuracy of daily DLW ME;
- -> underestimate of uphill walking, running V'O2 ME

measures



Global Positioning System

- Global Positioning System (GPS) is a space-based satellite navigation system that provides location and time information in all weather conditions, anywhere on or near the Earth where there is an unobstructed line of sight to four or more GPS satellites;

- provides critical capabilities to also commercial users around the world; - is maintained by the USA government and is freely accessible to anyone with a GPS receiver;







